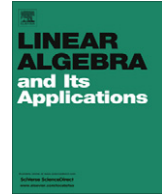




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Distance bounds for prescribed multiple eigenvalues of matrix polynomials

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ABSTRACT

In this paper, motivated by a problem posed by Wilkinson, we study the coefficient perturbations of a (square) matrix polynomial to a matrix polynomial that has a prescribed eigenvalue of specified algebraic multiplicity and index of annihilation. For an $n \times n$ matrix polynomial $P(\lambda)$ and a given scalar $\mu \in \mathbb{C}$, we introduce two weighted spectral norm distances, $\mathcal{E}_r(\mu)$ and $\mathcal{E}_{r,k}(\mu)$, from $P(\lambda)$ to the $n \times n$ matrix polynomials that have μ as an eigenvalue of algebraic multiplicity at least r and to those that have μ as an eigenvalue of algebraic multiplicity at least r and maximum Jordan chain length (exactly) k , respectively. Then we obtain a lower bound for $\mathcal{E}_{r,k}(\mu)$, and derive an upper bound for $\mathcal{E}_r(\mu)$ by constructing an associated perturbation of $P(\lambda)$.

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1. Introduction and preliminaries

Wilkinson's problem [14, 15] concerns computing the spectral norm distance (known as *Wilkinson's distance*) from an $n \times n$ matrix with n distinct eigenvalues to the set of $n \times n$ matrices having multiple eigenvalues, and has a strong connection to ill-conditioning of eigenvalue problems. Malyshev [9] provided a solution to Wilkinson's problem by obtaining a singular value characterization of Wilkinson's

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distance. Recently, Mengi [11] (extending Malyshev's methodology) derived a singular value optimization characterization for the smallest perturbation to a matrix that has an eigenvalue of specified algebraic multiplicity. Papathanasiou and Psarrakos [13] studied the case of polynomial eigenvalue problems, and applied Malyshev's technique to derive lower and upper bounds for a weighted distance from a given $n \times n$ matrix polynomial to the $n \times n$ matrix polynomials that have a prescribed multiple eigenvalue.

Motivated by the above, we consider an $n \times n$ matrix polynomial

$$P(\lambda) = A_m \lambda^m + A_{m-1} \lambda^{m-1} + \cdots + A_1 \lambda + A_0, \quad (1)$$

where λ is a complex variable and $A_j \in \mathbb{C}^{n \times n}$ ($j = 0, 1, \dots, m$) with $A_m \neq 0$. We also assume that $P(\lambda)$ is *regular*, that is, the determinant $\det P(\lambda)$ is not identically zero. The study of matrix polynomials, especially with regard to their spectral analysis, has a long history and important applications; see [4,8,10] and the references therein.

A scalar $\lambda_0 \in \mathbb{C}$ is called an *eigenvalue* of $P(\lambda)$ if the system $P(\lambda_0)x = 0$ has a nonzero solution $x_0 \in \mathbb{C}^n$. This solution x_0 is known as a (*right*) *eigenvector* of $P(\lambda)$ corresponding to λ_0 . The set of all eigenvalues of $P(\lambda)$, $\sigma(P) = \{\lambda \in \mathbb{C} : \det P(\lambda) = 0\}$, is the *spectrum* of $P(\lambda)$, and since $P(\lambda)$ is regular, it contains no more than nm finite elements. The *algebraic multiplicity* of a $\lambda_0 \in \sigma(P)$ is the multiplicity of λ_0 as a zero of the (scalar) polynomial $\det P(\lambda)$, and it is always greater than or equal to the *geometric multiplicity* of λ_0 , that is, the dimension of the null space of matrix $P(\lambda_0)$. An eigenvalue of $P(\lambda)$ is called *semisimple* if its algebraic and geometric multiplicities are equal, and it is called *defective* otherwise.

Suppose that for an eigenvalue $\lambda_0 \in \sigma(P)$, there exist $x_0, x_1, \dots, x_k \in \mathbb{C}^n$ with $x_0 \neq 0$, such that

$$\sum_{j=0}^{\xi} \frac{1}{j!} P^{(j)}(\lambda_0) x_{\xi-j} = 0; \quad \xi = 0, 1, \dots, k, \quad (2)$$

where $P^{(j)}(\lambda)$ denotes the j th derivative of $P(\lambda)$ and $k+1$ cannot exceed the algebraic multiplicity of λ_0 . Then the vector x_0 is clearly an eigenvector of λ_0 , and the vectors x_1, x_2, \dots, x_k are known as *generalized eigenvectors*. The set $\{x_0, x_1, \dots, x_k\}$ is called a *Jordan chain of length $k+1$* of $P(\lambda)$ corresponding to the eigenvalue λ_0 . Moreover, it is apparent that any set $\{x_0, x_1, \dots, x_{\xi}\}$, $\xi = 0, 1, \dots, k-1$, is also a Jordan chain of $P(\lambda)$ corresponding to λ_0 . Any eigenvalue of $P(\lambda)$ of geometric multiplicity p has p maximal Jordan chains associated with p linearly independent eigenvectors, with total number of eigenvectors and generalized eigenvectors equal to the algebraic multiplicity of this eigenvalue. The largest length of Jordan chains of $P(\lambda)$ corresponding to $\lambda_0 \in \sigma(P)$ is known as the *index of annihilation* of λ_0 [7]. This index coincides with the size of the largest Jordan blocks of the Jordan canonical form of $P(\lambda)$ corresponding to λ_0 , and it is equal to 1 if and only if the eigenvalue λ_0 is semisimple; for details on the Jordan structure of matrix polynomials, see [4,8].

We are interested in (additive) perturbations of the matrix polynomial $P(\lambda)$ of the form

$$Q(\lambda) = P(\lambda) + \Delta(\lambda) = \sum_{j=0}^m (A_j + \Delta_j) \lambda^j, \quad (3)$$

where the matrices $\Delta_j \in \mathbb{C}^{n \times n}$ ($j = 0, 1, \dots, m$) are arbitrary. For a given parameter $\varepsilon > 0$ and a given set of nonnegative weights $w = \{w_0, w_1, \dots, w_m\}$ with $w_0 > 0$, we define the class of admissible perturbed matrix polynomials

$$\mathcal{B}(P, \varepsilon, w) = \{Q(\lambda) \text{ as in (3)} : \|\Delta_j\| \leq \varepsilon w_j, j = 0, 1, \dots, m\},$$

where $\|\cdot\|$ denotes the *spectral matrix norm* (i.e., the norm subordinate to the Euclidean vector norm), and the polynomial $w(\lambda) = w_m \lambda^m + \cdots + w_1 \lambda + w_0$. The weights w_0, w_1, \dots, w_m allow freedom

in how perturbations are measured. Moreover, $\mathcal{B}(P, \varepsilon, w)$ is convex and compact, with respect to the max norm $\|P(\lambda)\|_\infty = \max_{0 \leq j \leq m} \|A_j\|$ [1].

Now we can introduce weighted distances from $P(\lambda)$ to the matrix polynomials that have a prescribed eigenvalue of algebraic multiplicity at least r , and to those that have a prescribed eigenvalue of algebraic multiplicity at least r and index of annihilation (exactly) k .

Definition 1.1 For the matrix polynomial $P(\lambda)$ in (1) and a given $\mu \in \mathbb{C}$, we define the distance from $P(\lambda)$ to μ as an eigenvalue of algebraic multiplicity at least r by

$$\mathcal{E}_r(\mu) = \inf \{ \varepsilon \geq 0 : \exists Q(\lambda) \in \mathcal{B}(P, \varepsilon, w) \text{ with } \mu \text{ as an eigenvalue of algebraic multiplicity at least } r \},$$

and the distance from $P(\lambda)$ to μ as an eigenvalue of algebraic multiplicity at least r and index of annihilation k by

$$\mathcal{E}_{r,k}(\mu) = \inf \{ \varepsilon \geq 0 : \exists Q(\lambda) \in \mathcal{B}(P, \varepsilon, w) \text{ with } \mu \text{ as an eigenvalue of algebraic multiplicity at least } r \text{ and index of annihilation } k \}.$$

Notice that we have the identity $\mathcal{E}_r(\mu) = \min_{k=1,2,\dots,mn} \mathcal{E}_{r,k}(\mu)$. Here, we allow values of k greater than r because the optimal perturbed matrix polynomial $Q(\lambda)$ may have μ as an eigenvalue of (both) algebraic multiplicity and index of annihilation greater than r .

The singular values of an $n \times n$ complex matrix A , i.e., the nonnegative square roots of the eigenvalues of A^*A , are denoted by $s_1(A) \geq s_2(A) \geq \dots \geq s_n(A) \geq 0$. It is apparent that for the matrix polynomial $P(\lambda)$, $\sigma(P) = \{\lambda \in \mathbb{C} : s_n(P(\lambda)) = 0\}$, and a scalar $\lambda_0 \in \mathbb{C}$ is an eigenvalue of $P(\lambda)$ of geometric multiplicity exactly p if and only if $s_1(P(\lambda_0)) \geq \dots \geq s_{n-p}(P(\lambda_0)) > s_{n-p+1}(P(\lambda_0)) = \dots = s_n(P(\lambda_0)) = 0$.

If $P(\lambda) = I\lambda - A$ for some $A \in \mathbb{C}^{n \times n}$, then $\sigma(P)$ coincides with the standard spectrum of A , $\sigma(A)$, and if in addition $w = \{w_0, w_1\} = \{1, 0\}$, then $\mathcal{B}(P, \varepsilon, w) = \{I\lambda - (A + E) : \|E\| \leq \varepsilon\}$. In this case, Malyshev [9] (inspired by [12]) has proved that

$$\mathcal{E}_2(\mu) = \sup_{\gamma > 0} s_{2n-1} \left(\begin{bmatrix} I\mu - A & 0 \\ \gamma I & I\mu - A \end{bmatrix} \right).$$

Extending Malyshev's methodology, Ikramov and Nazari [6] (for $r = 3$), and Mengi [11] have obtained (always for $P(\lambda) = I\lambda - A$ and $w = \{w_0, w_1\} = \{1, 0\}$, i.e., for constant matrices)

$$\mathcal{E}_r(\mu) = \sup_{\gamma_{i,j} \in \mathbb{C} \setminus \{0\}} s_{rn-r+1} \left(\begin{bmatrix} I\mu - A & 0 & 0 & \dots & 0 \\ \gamma_{2,1}I & I\mu - A & 0 & \dots & 0 \\ \gamma_{3,1}I & \gamma_{3,2}I & I\mu - A & \dots & 0 \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ \gamma_{r,1}I & \gamma_{r,2}I & \gamma_{r,3}I & \dots & I\mu - A \end{bmatrix} \right). \quad (4)$$

In this work, our goal is to derive a primer estimation of the distances $\mathcal{E}_{r,k}(\mu)$ and $\mathcal{E}_r(\mu)$ for matrix polynomials. Unfortunately, in the case of matrix polynomials, Malyshev's technique leads to lower and upper bounds for the distance $\mathcal{E}_2(\mu)$ and not to its exact value [13]. Hence, to be able to exploit the definition of Jordan chains of matrix polynomials in (2) and avoid multivariable optimization problems, we consider block-Toeplitz matrices of higher order with only one real parameter $\gamma > 0$

and its powers instead of the $r(r-1)/2$ complex parameters $\gamma_{2,1}, \gamma_{3,1}, \gamma_{3,2}, \dots, \gamma_{r,r-1}$ in (4) (see Definition 2.1 below and the discussion before Theorem 2.4). Generalizing results of [13], in Section 2, we obtain a lower bound for $\varepsilon_{r,k}(\mu)$, and in Section 3, we construct an upper bound for $\varepsilon_r(\mu)$ and an associated perturbation of $P(\lambda)$. Simple numerical examples are given in Section 4 to illustrate our results.

2. A lower bound for the distance $\varepsilon_{r,k}(\mu)$

By Theorem 4 of [13], we know that if both the required algebraic and geometric multiplicities of μ are equal to r (i.e., the index of annihilation is equal to 1), then $\varepsilon_{r,1}(\mu) \geq s_{n-r+1}(P(\mu))/w(|\mu|)$. Hence, it follows that

$$\varepsilon_1(\mu) = \frac{s_n(P(\mu))}{w(|\mu|)} \leq \varepsilon_r(\mu) \leq \frac{s_{n-r+1}(P(\mu))}{w(|\mu|)} \leq \varepsilon_{r,1}(\mu).$$

So, in the special case $s_n(P(\mu)) = s_{n-1}(P(\mu)) = \dots = s_{n-r+1}(P(\mu))$, it is clear that $\varepsilon_r(\mu) = s_{n-r+1}(P(\mu))/w(|\mu|)$, and an optimal perturbation of $P(\lambda)$ is given by [13, Formula (3)]. Thus, in what follows, we assume that $s_n(P(\mu)) \neq s_{n-r+1}(P(\mu))$ and consider perturbations of $P(\lambda)$ such that the perturbed matrix polynomial has μ as a defective eigenvalue (i.e., $k \geq 2$) of algebraic multiplicity at least r . The next definition is necessary for the remainder.

Definition 2.1 For the matrix polynomial $P(\lambda)$ in (1), a positive integer k and a scalar $\gamma \in \mathbb{C}$, we define the $kn \times kn$ matrix polynomial

$$F_k[P(\lambda); \gamma] = \begin{bmatrix} P(\lambda) & 0 & \cdots & 0 \\ \gamma P^{(1)}(\lambda) & P(\lambda) & \cdots & 0 \\ \frac{\gamma^2}{2!} P^{(2)}(\lambda) & \gamma P^{(1)}(\lambda) & \cdots & 0 \\ \vdots & \vdots & \ddots & \vdots \\ \frac{\gamma^{k-1}}{(k-1)!} P^{(k-1)}(\lambda) & \frac{\gamma^{k-2}}{(k-2)!} P^{(k-2)}(\lambda) & \cdots & P(\lambda) \end{bmatrix}.$$

We observe that a scalar $\lambda_0 \in \mathbb{C}$ is an eigenvalue of $P(\lambda)$ if and only if it is an eigenvalue of $F_k[P(\lambda); \gamma]$. Furthermore, if λ_0 is an eigenvalue of $P(\lambda)$ of algebraic multiplicity r and index of annihilation k , then for every $\gamma \neq 0$, the null space of matrix $F_k[P(\lambda_0); \gamma]$ has dimension at least r (for $\gamma = 1$, see [5, Lemma 2.5] and [7]).

Lemma 2.2 Suppose $\lambda_0 \in \mathbb{C}$ is an eigenvalue of $P(\lambda)$ of algebraic multiplicity at least r and index of annihilation k . Then for any nonzero $\gamma \in \mathbb{C}$, we have $s_{kn-r+1}(F_k[P(\lambda_0); \gamma]) = 0$.

Proof. By hypothesis, we may assume that $P(\lambda)$ has p Jordan chains corresponding to λ_0 , namely,

$$\{x_{1,0}, x_{1,1}, \dots, x_{1,k_1}\}, \{x_{2,0}, x_{2,1}, \dots, x_{2,k_2}\}, \dots, \{x_{p,0}, x_{p,1}, \dots, x_{p,k_p}\},$$

with $x_{1,0}, x_{2,0}, \dots, x_{p,0}$ linearly independent eigenvectors, $k-1 = k_1 \geq k_2 \geq \dots \geq k_p$ and $(k_1+1) + (k_2+1) + \dots + (k_p+1) = r$. Notice that the first Jordan chain, $\{x_{1,0}, x_{1,1}, \dots, x_{1,k_1}\}$, is necessarily maximal. Then, recalling (2), we see that the r vectors

$$\begin{bmatrix} 0 \\ \vdots \\ 0 \\ x_{i,0} \\ \gamma x_{i,1} \\ \gamma^2 x_{i,2} \\ \vdots \\ \gamma^{k_i-1} x_{i,k_i-1} \\ \gamma^{k_i} x_{i,k_i} \end{bmatrix}, \begin{bmatrix} 0 \\ \vdots \\ 0 \\ x_{i,0} \\ \gamma x_{i,1} \\ \vdots \\ \gamma^{k_i-2} x_{i,k_i-2} \\ \gamma^{k_i-1} x_{i,k_i-1} \end{bmatrix}, \dots, \begin{bmatrix} 0 \\ \vdots \\ 0 \\ 0 \\ 0 \\ 0 \\ \vdots \\ 0 \\ 0 \\ x_{i,0} \end{bmatrix} \in \mathbb{C}^{kn}; \quad i = 1, 2, \dots, p \quad (5)$$

satisfy

$$F_k[P(\lambda_0); \gamma] \begin{bmatrix} 0 \\ \vdots \\ 0 \\ x_{i,0} \\ \gamma x_{i,1} \\ \gamma^2 x_{i,2} \\ \vdots \\ \gamma^{\xi-1} x_{i,\xi-1} \\ \gamma^{\xi} x_{i,\xi} \end{bmatrix} = \begin{bmatrix} 0 \\ \vdots \\ 0 \\ P(\lambda_0) x_{i,0} \\ \gamma \sum_{j=0}^1 \frac{1}{j!} P^{(j)}(\lambda_0) x_{i,1-j} \\ \gamma^2 \sum_{j=0}^2 \frac{1}{j!} P^{(j)}(\lambda_0) x_{i,2-j} \\ \vdots \\ \gamma^{\xi} \sum_{j=0}^{\xi} \frac{1}{j!} P^{(j)}(\lambda_0) x_{i,\xi-j} \end{bmatrix} = 0$$

for all $i = 1, 2, \dots, p$ and $\xi = 0, 1, \dots, k_i$. Hence, they lie in the null space of matrix $F_k[P(\lambda_0); \gamma]$. Moreover, one can verify that the vectors in (5) are linearly independent, keeping in mind their block form and the linear independence of the eigenvectors $x_{1,0}, x_{2,0}, \dots, x_{p,0} \in \mathbb{C}^n$. As a consequence, the dimension of the null space of $F_k[P(\lambda_0); \gamma]$ is greater than or equal to r (i.e., 0 is an eigenvalue of matrix $F_k[P(\lambda_0); \gamma]$ of geometric multiplicity at least r), and thus, $s_{kn-r+1}(F_k[P(\lambda_0); \gamma]) = 0$. \square

By this lemma, if a scalar $\mu \in \mathbb{C}$ is a multiple eigenvalue of a perturbed matrix polynomial $Q(\lambda) = P(\lambda) + \Delta(\lambda)$ (as in (3)) of algebraic multiplicity at least r and index of annihilation k , then for any nonzero $\gamma \in \mathbb{C}$, μ is an eigenvalue of the $kn \times kn$ matrix polynomial $F_k[Q(\lambda); \gamma]$ of geometric multiplicity at least r . This observation and the discussion in Section 3 of [13] yield readily the following lemma.

Lemma 2.3 *If $\mu \in \mathbb{C}$ is an eigenvalue of a matrix polynomial $Q(\lambda) = P(\lambda) + \Delta(\lambda)$ of algebraic multiplicity at least r and index of annihilation k , then for every $\gamma \neq 0$,*

$$s_{kn-r+1}(F_k[P(\mu); \gamma]) \leq \|F_k[\Delta(\mu); \gamma]\|.$$

It is straightforward to verify that $\begin{bmatrix} u_1 \\ u_2 \\ \vdots \\ u_k \end{bmatrix}, \begin{bmatrix} v_1 \\ v_2 \\ \vdots \\ v_k \end{bmatrix} \in \mathbb{C}^{kn} (u_j, v_j \in \mathbb{C}^n, j = 1, 2, \dots, k)$ is a pair of

left and right singular vectors corresponding to a singular value of $F_k[P(\mu); \gamma]$ ($\gamma \neq 0$) if and only

if $\begin{bmatrix} u_1 \\ (\bar{\gamma}/|\gamma|)u_2 \\ \vdots \\ (\bar{\gamma}/|\gamma|)^{k-1}u_k \end{bmatrix}, \begin{bmatrix} v_1 \\ (\bar{\gamma}/|\gamma|)v_2 \\ \vdots \\ (\bar{\gamma}/|\gamma|)^{k-1}v_k \end{bmatrix}$ is a pair of left and right singular vectors of $F_k[P(\mu); |\gamma|]$

corresponding to the same singular value. Hence, for convenience, and without loss of generality, from this point and in the remainder of the paper, we assume that the parameter γ is real positive.

Theorem 2.4 Suppose $\mu \in \mathbb{C}$ is an eigenvalue of a perturbed matrix polynomial $Q(\lambda) = P(\lambda) + \Delta(\lambda) \in \mathcal{B}(P, \varepsilon, w)$ of algebraic multiplicity at least r and index of annihilation k . Then for every $\gamma > 0$,

$$\varepsilon \geq \frac{\|F_k[\Delta(\mu); \gamma]\|}{\|F_k[w(|\mu|); |\gamma|]\|} \geq \frac{s_{kn-r+1}(F_k[P(\mu); \gamma])}{\|F_k[w(|\mu|); |\gamma|]\|}.$$

Proof. For the matrix polynomial $\Delta(\lambda)$ and its derivatives, we have

$$\|\Delta(\mu)\| \leq \sum_{j=0}^m \|\Delta_j\| |\mu|^j \leq \varepsilon w(|\mu|)$$

and

$$\|\Delta^{(i)}(\mu)\| \leq \sum_{j=i}^m j(j-1)\cdots(j-i+1) \|\Delta_j\| |\mu|^{j-i} \leq \varepsilon w^{(i)}(|\mu|); \quad i = 1, 2, \dots, r-1.$$

Thus, for any $\gamma > 0$, there is a unit vector $\hat{y} = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_k \end{bmatrix} \in \mathbb{C}^{kn}$ with $y_1, y_2, \dots, y_k \in \mathbb{C}^n$ such that

$$\begin{aligned} \|F_k[\Delta(\mu); \gamma]\|^2 &= \|F_k[\Delta(\mu); \gamma]\hat{y}\|^2 = \left\| \begin{bmatrix} \Delta(\mu)y_1 \\ \gamma \Delta^{(1)}(\mu)y_1 + \Delta(\mu)y_2 \\ \vdots \\ \sum_{i=0}^{k-1} \frac{\gamma^i}{i!} \Delta^{(i)}(\mu)y_{k-i} \end{bmatrix} \right\|^2 \\ &= \|\Delta(\mu)y_1\|^2 + \|\gamma \Delta^{(1)}(\mu)y_1 + \Delta(\mu)y_2\|^2 + \cdots + \left\| \sum_{i=0}^{k-1} \frac{\gamma^i}{i!} \Delta^{(i)}(\mu)y_{k-i} \right\|^2 \end{aligned}$$

$$\begin{aligned}
&\leq \|\Delta(\mu)\|^2 \|y_1\|^2 + \gamma^2 \|\Delta^{(1)}(\mu)\|^2 \|y_1\|^2 \\
&\quad + 2\gamma \|\Delta(\mu)\| \|\Delta^{(1)}(\mu)\| \|y_1\| \|y_2\| + \|\Delta(\mu)\|^2 \|y_2\|^2 \\
&\quad + \cdots + \|\Delta(\mu)\|^2 \|y_k\|^2 \\
&\leq (\varepsilon w(|\mu|))^2 \|y_1\|^2 + \gamma^2 (\varepsilon w^{(1)}(|\mu|))^2 \|y_1\|^2 \\
&\quad + 2\gamma (\varepsilon w(|\mu|)) (\varepsilon w^{(1)}(|\mu|)) \|y_1\| \|y_2\| + (\varepsilon w(|\mu|))^2 \|y_2\|^2 \\
&\quad + \cdots + (\varepsilon w(|\mu|))^2 \|y_k\|^2 \\
&= \left\| \begin{bmatrix} \varepsilon w(|\mu|) \|y_1\| \\ \gamma \varepsilon w^{(1)}(|\mu|) \|y_1\| + \varepsilon w(|\mu|) \|y_2\| \\ \vdots \\ \sum_{i=0}^{k-1} \frac{\gamma^i}{i!} \varepsilon w^{(i)}(|\mu|) \|y_{k-i}\| \end{bmatrix} \right\|^2 \\
&\leq \varepsilon^2 \|F_k[w(|\mu|); \gamma]\|^2.
\end{aligned}$$

The proof is completed by Lemma 2.3. \square

As mentioned before, for $k = 1$, $\mathcal{E}_{r,1}(\mu) \geq s_{n-r+1}(P(\mu))/w(|\mu|)$ [13, Theorem 4].

Corollary 2.5 For any $\gamma > 0$, the inequalities

$$\mathcal{E}_{r,k}(\mu) \geq \frac{s_{kn-r+1}(F_k[P(\mu); \gamma])}{\|F_k[w(|\mu|); \gamma]\|}; \quad k = 1, 2, \dots, r$$

and

$$\mathcal{E}_r(\mu) \geq \min_{k=1,2,\dots,nm} \frac{s_{kn-r+1}(F_k[P(\mu); \gamma])}{\|F_k[w(|\mu|); \gamma]\|}$$

hold.

Let us denote by $\eta(r, k)$ the smallest integer that is greater than or equal to $n - \frac{r-1}{k}$. One can see that

$$\frac{s_{kn-r+1}(F_k[P(\mu); \gamma])}{\|F_k[w(|\mu|); \gamma]\|} \longrightarrow \frac{s_{\eta(r,k)}(P(\mu))}{w(|\mu|)} \leq \mathcal{E}_{n-\eta(r,k)+1,1}(\mu)$$

as $\gamma \longrightarrow 0^+$.

3. An upper bound for the distance $\mathcal{E}_r(\mu)$

The technique applied in the proof of Theorem 11 of [13] can be extended for the derivation of an upper bound for the distance $\mathcal{E}_r(\mu)$ from $P(\lambda)$ in (1) to the $n \times n$ matrix polynomials that have $\mu \in \mathbb{C}$ as an eigenvalue of algebraic multiplicity at least r . The following definition is necessary.

Definition 3.1 For any $\gamma > 0$ and $r \in \{2, 3, \dots, n\}$, let
$$\begin{bmatrix} u_1(\gamma) \\ u_2(\gamma) \\ \vdots \\ u_r(\gamma) \end{bmatrix}, \begin{bmatrix} v_1(\gamma) \\ v_2(\gamma) \\ \vdots \\ v_r(\gamma) \end{bmatrix} \in \mathbb{C}^m(u_j(\gamma), v_j(\gamma) \in \mathbb{C}^n, j = 1, 2, \dots, r)$$

be a pair of left and right singular vectors of $s_{m-r+1}(F_r[P(\mu); \gamma])$, respectively. Then we define the $n \times r$ matrices

$$U(\gamma) = [u_1(\gamma) \ u_2(\gamma) \ \cdots \ u_r(\gamma)] \quad \text{and} \quad V(\gamma) = [v_1(\gamma) \ v_2(\gamma) \ \cdots \ v_r(\gamma)].$$

For any $\gamma > 0$ with $\text{rank}(V(\gamma)) = r \in \{2, 3, \dots, n\}$, we will construct a perturbation $\Delta_\gamma(\lambda)$ such that the perturbed matrix polynomial $Q_\gamma(\lambda) = P(\lambda) + \Delta_\gamma(\lambda)$ has μ as a defective eigenvalue with an associated (not necessarily maximal) Jordan chain of length r . First we define the quantities

$$\phi_i = \frac{w^{(i)}(|\mu|)}{(i!) w(|\mu|)} \left(\frac{\bar{\mu}}{|\mu|} \right)^i; \quad i = 1, 2, \dots, r,$$

setting $\bar{\mu}/|\mu| = 0$ whenever $\mu = 0$, and recalling that $w_0 > 0$. We also consider the $r \times r$ upper triangular Toeplitz matrix

$$\Theta_\gamma = [\theta_{i,j}] = \begin{bmatrix} 1 - \gamma\phi_1 & \gamma^2(\phi_1^2 - \phi_2) & \gamma^3(2\phi_1\phi_2 - \phi_3 - \phi_1^3) & \cdots \\ 0 & 1 & -\gamma\phi_1 & \gamma^2(\phi_1^2 - \phi_2) & \cdots \\ 0 & 0 & 1 & -\gamma\phi_1 & \cdots \\ 0 & 0 & 0 & 1 & \cdots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix},$$

whose entries above the main diagonal are given by the recursive formulae

$$\theta_{i,j} = -\theta_{i,i}\gamma^{j-i}\phi_{j-i} - \theta_{i,i+1}\gamma^{j-(i+1)}\phi_{j-(i+1)} - \cdots - \theta_{i,j-1}\gamma\phi_1; \quad 1 \leq i < j \leq r. \quad (6)$$

Denoting by $V(\gamma)^\dagger$ the Moore–Penrose pseudoinverse of $V(\gamma)$, we consider the $n \times n$ matrix

$$\Delta_\gamma = -s_{m-r+1}(F_r[P(\mu); \gamma]) U(\gamma) \Theta_\gamma V(\gamma)^\dagger,$$

and define the matrices

$$\Delta_{\gamma,j} = \frac{w_j}{w(|\mu|)} \left(\frac{\bar{\mu}}{|\mu|} \right)^j \Delta_\gamma; \quad j = 0, 1, \dots, m$$

and the matrix polynomial

$$\Delta_\gamma(\lambda) = \sum_{j=0}^m \Delta_{\gamma,j} \lambda^j.$$

We observe that $\Delta_\gamma(\mu) = \Delta_\gamma$, and for $i = 1, 2, \dots, r$,

$$\begin{aligned}
\Delta_{\gamma}^{(i)}(\mu) &= \sum_{j=i}^m j(j-1)\cdots(j-i+1) \frac{w_j}{w(|\mu|)} \left(\frac{\bar{\mu}}{|\mu|}\right)^j \Delta_{\gamma} \mu^{j-i} \\
&= \Delta_{\gamma} \frac{1}{w(|\mu|)} \sum_{j=i}^m j(j-1)\cdots(j-i+1) w_j \left(\frac{\bar{\mu}}{|\mu|}\right)^i \left(\frac{\bar{\mu}}{|\mu|}\right)^{j-i} \mu^{j-i} \\
&= \Delta_{\gamma} \frac{w^{(i)}(|\mu|)}{w(|\mu|)} \left(\frac{\bar{\mu}}{|\mu|}\right)^i = (i!) \phi_i \Delta_{\gamma}.
\end{aligned}$$

Since $u(\gamma), v(\gamma) \in \mathbb{C}^m$ are a left and a right singular vector of $s_{m-r+1}(F_r[P(\mu); \gamma])$, respectively, it holds that

$$F_r[P(\mu); \gamma] v(\gamma) = s_{m-r+1}(F_r[P(\mu); \gamma]) u(\gamma). \quad (7)$$

Denote also by e_1, e_2, \dots the vectors of the standard basis, i.e., the columns of the identity matrix. If $\text{rank}(V(\gamma)) = r$ ($\in \{2, 3, \dots, n\}$), or equivalently, if $V(\gamma)^{\dagger} V(\gamma) = I_r$ (the $r \times r$ identity matrix), then the perturbed matrix polynomial

$$Q_{\gamma}(\lambda) = P(\lambda) + \Delta_{\gamma}(\lambda) = \sum_{j=0}^m (A_j + \Delta_{\gamma,j}) \lambda^j \quad (8)$$

satisfies

$$\begin{aligned}
Q_{\gamma}(\mu) v_1(\gamma) &= P(\mu) v_1(\gamma) + \Delta_{\gamma} v_1(\gamma) \\
&= s_{m-r+1}(F_r[P(\mu); \gamma]) u_1(\gamma) - s_{m-r+1}(F_r[P(\mu); \gamma]) U(\gamma) e_1 \\
&= s_{m-r+1}(F_r[P(\mu); \gamma]) u_1(\gamma) - s_{m-r+1}(F_r[P(\mu); \gamma]) u_1(\gamma) \\
&= 0.
\end{aligned}$$

By straightforward computations, setting $\phi_0 = 1$ and recalling (6) and (7), we verify that for every $i = 1, 2, \dots, r-1$,

$$\begin{aligned}
&\sum_{j=0}^i \frac{1}{j!} \gamma^j Q_{\gamma}^{(j)}(\mu) v_{i-j+1}(\gamma) \\
&= \sum_{j=0}^i \frac{1}{j!} \gamma^j P^{(j)}(\mu) v_{i-j+1}(\gamma) + \sum_{j=0}^i \frac{1}{j!} \gamma^j \Delta_{\gamma}^{(j)}(\mu) v_{i-j+1}(\gamma) \\
&= s_{m-r+1}(F_r[P(\mu); \gamma]) u_{i+1}(\gamma) - \sum_{j=0}^i \gamma^j \phi_j \Delta_{\gamma} v_{i-j+1}(\gamma) \\
&= s_{m-r+1}(F_r[P(\mu); \gamma]) \left(u_{i+1}(\gamma) - \sum_{j=0}^i \gamma^j \phi_j U(\gamma) \Theta_{\gamma} V(\gamma)^{\dagger} v_{i-j+1}(\gamma) \right) \\
&= s_{m-r+1}(F_r[P(\mu); \gamma]) \left(u_{i+1}(\gamma) - \sum_{j=0}^i \gamma^j \phi_j U(\gamma) \Theta_{\gamma} e_{i-j+1} \right) \\
&= s_{m-r+1}(F_r[P(\mu); \gamma]) \left(u_{i+1}(\gamma) - \sum_{j=0}^i \sum_{\xi=1}^{i-j+1} \gamma^j \phi_j \theta_{\xi, i-j+1} u_{\xi}(\gamma) \right)
\end{aligned}$$

$$\begin{aligned}
&= s_{m-r+1}(F_r[P(\mu); \gamma]) \left(u_{i+1}(\gamma) - \theta_{i+1, i+1} u_{i+1}(\gamma) - \sum_{\xi=1}^i \left(\sum_{j=\xi}^{i+1} \gamma^{i-j+1} \phi_{i-j+1} \theta_{\xi, j} \right) u_{\xi}(\gamma) \right) \\
&= -s_{m-r+1}(F_r[P(\mu); \gamma]) \left(\sum_{\xi=1}^i \left(\sum_{j=\xi}^{i+1} \gamma^{i-j+1} \phi_{i-j+1} \theta_{\xi, j} \right) u_{\xi}(\gamma) \right) \\
&= 0.
\end{aligned}$$

Dividing by $\gamma^i \neq 0$ yields

$$\sum_{j=0}^i \frac{1}{j!} Q_{\gamma}^{(j)}(\mu) \left(\gamma^{-(i-j)} v_{i-j+1}(\gamma) \right) = 0.$$

As a consequence, if $\text{rank}(V(\gamma)) = r$ ($\in \{2, 3, \dots, n\}$), then μ is a defective eigenvalue of $Q_{\gamma}(\lambda)$ with $\{v_1(\gamma), \gamma^{-1}v_2(\gamma), \gamma^{-2}v_3(\gamma), \dots, \gamma^{-(r-1)}v_r(\gamma)\}$ as an associated Jordan chain of length r (recall the definition of Jordan chains in (2)).

Furthermore, we see that

$$\|\Delta_{\gamma}(\mu)\| = \|\Delta_{\gamma}\| = s_{m-r+1}(F_r[P(\mu); \gamma]) \|U(\gamma) \Theta_{\gamma} V(\gamma)^{\dagger}\|$$

and

$$\|\Delta_{\gamma, j}\| = w_j \frac{s_{m-r+1}(F_r[P(\mu); \gamma])}{w(|\mu|)} \|U(\gamma) \Theta_{\gamma} V(\gamma)^{\dagger}\|; \quad j = 0, 1, \dots, r.$$

Hence, we have the next result, which is a direct generalization of the second part of Theorem 11 in [13].

Theorem 3.2 Let $P(\lambda)$ be a matrix polynomial as in (1), $\mu \in \mathbb{C}$, and $r \in \{2, 3, \dots, n\}$. Then for every $\gamma > 0$ such that $\text{rank}(V(\gamma)) = r$, it holds that

$$\varepsilon_r(\mu) \leq \frac{s_{m-r+1}(F_r[P(\mu); \gamma])}{w(|\mu|)} \|U(\gamma) \Theta_{\gamma} V(\gamma)^{\dagger}\|,$$

and $Q_{\gamma}(\lambda)$ in (8) lies on the boundary of $\mathcal{B}\left(P, \frac{s_{m-r+1}(F_r[P(\mu); \gamma])}{w(|\mu|)} \|U(\gamma) \Theta_{\gamma} V(\gamma)^{\dagger}\|, w\right)$ and has μ as a defective eigenvalue with a (not necessarily maximal) Jordan chain of length r .

We remark that for $r \geq 3$, it is not easy to find values of γ for which the condition “ $\text{rank}(V(\gamma)) = r$ ” is ensured, as it was done in [13] for $r = 2$. On the other hand, in all our experiments, this rank condition appears to hold generically.

4. Numerical examples

To illustrate the proposed (lower and upper) bounds and their tightness, we begin with the special case of constant matrices.

Example 4.1 Consider the 6×6 smoke matrix that can be generated by the Matlab command `gallery('smoke', 6)`, the corresponding linear pencil $L(\lambda) = I_6\lambda - S$, the weights $w = \{w_0, w_1\} = \{1, 0\}$, and the scalar $\mu = 0.3841 + i0.6767$. By [11], we know that

$$\varepsilon_3(0.3841 + i0.6767) = \varepsilon_{3,3}(0.3841 + i0.6767) = 0.3270.$$

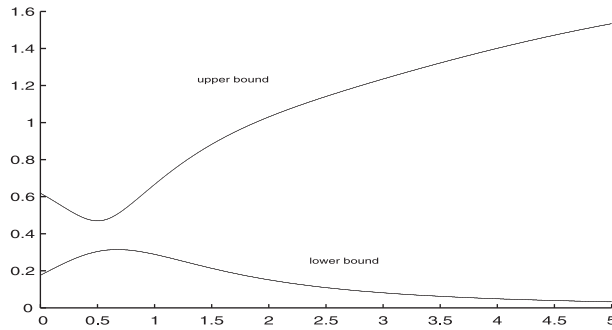


Fig. 1. Lower and upper bounds for $\varepsilon_3(0.3841 + i 0.6767)$, $0 < \gamma \leq 5$.

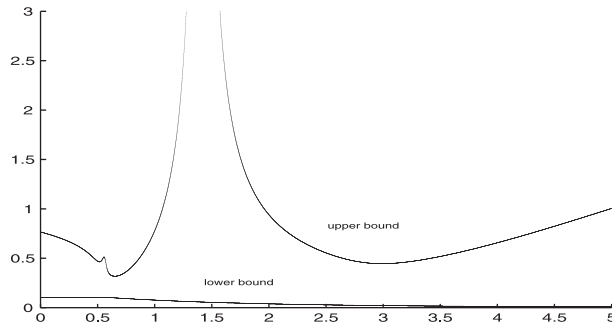


Fig. 2. Lower bound for $\varepsilon_{3,3}(-1.1105)$ and upper bound for $\varepsilon_3(-1.1105)$, $0 < \gamma \leq 5$.

The graphs of our lower bound for $\varepsilon_{3,3}(0.3841 + i 0.6767)$ (see Corollary 2.5) and our upper bound for $\varepsilon_3(0.3841 + i 0.6767)$ (see Theorem 3.2) are depicted in Fig. 1 for $\gamma \in (0, 5]$. For $\gamma = 0.6748$, Corollary 2.5 implies the lower bound 0.3145. Moreover, for $\gamma = 0.5004$, Theorem 3.2 yields the upper bound 0.4694 and an associated perturbed matrix $S + \Delta$ with $\|\Delta\| = 0.4694$, which has $\mu = 0.3841 + i 0.6767$ as a defective eigenvalue of algebraic multiplicity 3 and geometric multiplicity 1.

For our second example, we consider a real quadratic matrix polynomial.

Example 4.2 Let

$$P(\lambda) = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix} \lambda^2 + \begin{bmatrix} 0 & 1 & 0 \\ 0 & 3 & 1 \\ 0 & -1 & 6 \end{bmatrix} \lambda + \begin{bmatrix} 2 & 1 & 0 \\ -1 & 3 & 0 \\ 0 & 0 & 10 \end{bmatrix}$$

and $w = \{w_0, w_1, w_2\} = \{10, 6.1108, 3\}$ (the norms of the coefficient matrices). For the scalar $\mu = -1.1105$, by Example 2 of [13] (see also Proposition 17 and Theorem 18 in [1]), we know that

$$\varepsilon_2(-1.1105) = \varepsilon_{2,2}(-1.1105) = \varepsilon_1(-1.1105) = 0.1002. \quad (9)$$

The graphs of our lower bound for $\varepsilon_{3,3}(-1.1105)$ and our upper bound for $\varepsilon_3(-1.1105)$ are illustrated in Fig. 2 for $\gamma \in (0, 5]$. Setting $\gamma = 0.5530$ and $\gamma = 0.6518$, we get the lower bound 0.1048 and the upper bound 0.3177, respectively. It is worth noting that these bounds are clearly compatible with (9). Furthermore, Theorem 3.2 yields the perturbed matrix polynomial

$$Q(\lambda) = \begin{bmatrix} 0.5323 & -0.0453 & -0.6286 \\ 0.1664 & 1.6410 & 0.4567 \\ -0.0043 & -0.0977 & 2.7268 \end{bmatrix} \lambda^2 + \begin{bmatrix} 0.9526 & 1.0922 & 1.2805 \\ -0.3389 & 3.7312 & 0.0696 \\ 0.0087 & -0.8009 & 6.5565 \end{bmatrix} \lambda + \begin{bmatrix} 0.4411 & 0.8490 & -2.0955 \\ -0.4454 & 1.8035 & 1.5225 \\ -0.0142 & -0.3258 & 9.0893 \end{bmatrix}$$

that lies on the boundary of $B(P, 0.3177, w)$, and has $\mu = -1.1105$ as a defective eigenvalue of

algebraic multiplicity 3 and geometric multiplicity 1, with an associated eigenvector $x = \begin{bmatrix} 0.9571 \\ 0.2890 \\ -0.0194 \end{bmatrix}$.

The matrix polynomial in our last example is triangular, and hence, we can directly compute a (non-optimal) perturbation with the desired properties.

Example 4.3 Consider the 4×4 quadratic matrix polynomial

$$P(\lambda) = \begin{bmatrix} 2\lambda^2 - 5\lambda + 4 & 0 & 2\lambda + 1 & -\lambda + 6 \\ 0 & \lambda^2 + 2\lambda - 5 & -\lambda^2 & 0 \\ 0 & 0 & 2\lambda^2 - i8 & i\lambda \\ 0 & 0 & 0 & \lambda^2 - \lambda + 15 \end{bmatrix}$$

and the weights $w = \{w_0, w_1, w_2\} = \{1, 1, 1\}$. The graphs of the proposed lower bound for $\varepsilon_{3,2}(2)$ and upper bound for $\varepsilon_3(2)$ are plotted in Fig. 3 for $\gamma \in (0, 5]$. As $\gamma \rightarrow 0^+$ and for $\gamma = 1.1827$, we get the lower bound 0.4093 for $\varepsilon_{3,2}(2)$ and the upper bound 1.6357 for $\varepsilon_3(2)$, respectively. The exact values of the distances $\varepsilon_{3,2}(2)$ and $\varepsilon_3(2)$ are not known, but these bounds are consistent with the fact that the matrix polynomial

$$R(\lambda) = \begin{bmatrix} \lambda^2 - 4\lambda + 4 & 0 & 2\lambda + 1 & -\lambda + 6 \\ 0 & \lambda^2 + \lambda - 6 & -\lambda^2 & 0 \\ 0 & 0 & \lambda^2 - i9 & i\lambda \\ 0 & 0 & 0 & \lambda^2 + 16 \end{bmatrix}$$

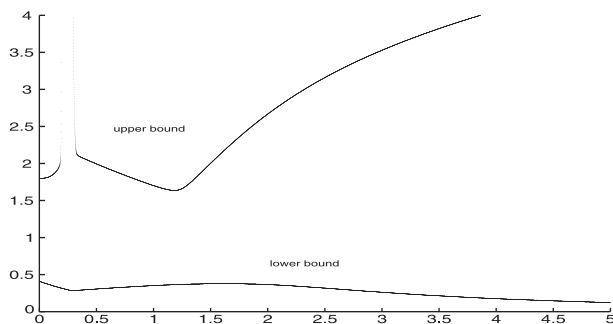


Fig. 3. Lower bound for $\varepsilon_{3,2}(2)$ and upper bound for $\varepsilon_3(2)$, $0 < \gamma \leq 5$.

lies on the boundary of $\mathcal{B}(P, 1, w)$, and has $\mu = 2$ as a defective eigenvalue of algebraic multiplicity 3 and index of annihilation 2. Recalling the comment in the last paragraph of Section 2 (and applying the methodology in [13, Section 3]), we also observe that the lower bound 0.4093 coincides with the distance $\mathcal{E}_{2,1}(2)$.

Corollary 2.5 and Theorem 3.2 provide an infinite number of lower bounds of $\mathcal{E}_{r,k}(\mu)$ and upper bounds of $\mathcal{E}_r(\mu)$, respectively, in such a way that the best lower/upper bound is given by the maximum/minimum of these bounds over all $\gamma > 0$. Since these optimization problems are over only one real variable, in our examples above, we apply the standard grid search with respect to $\gamma \in (0, 5]$. The difficulty with this brute-force approach is that it usually requires too many bound evaluations.

Alternatively, we can use golden section search [3, pp. 656–659], or Brent's method [2, Chapter 5]. The latter algorithm is based on the combination of inverse quadratic interpolation and golden section search, and one of its implementations is the Matlab function `fminbnd`. As it is expected, these two derivative-free optimization methods compute at best a local extremum, and there is no guarantee that the global extremum will be found (so, the grid search with a few grid points can be used to initiate an interval of the global optimum). For example, we consider the upper bound of the distance $\mathcal{E}_3(-1.1105)$ in Example 2 (see Fig. 2). Applying `fminbnd` on the interval $(0, 2]$, we get the global minimum 0.3177 at $\gamma = 0.6518$ after 8 iterations. Over the interval $(0, 5]$, `fminbnd` finds the local minimum 0.4449 at $\gamma = 2.9905$ after 10 iterations.

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